Title: Semantic Representation of Neural Circuit Knowledge in *Caenorhabditis elegans*

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Abstract

In modern biology, new knowledge is generated quickly, making it challenging for researchers to efficiently acquire and synthesise new information from the large volume of primary publications. To address this problem, computational approaches that generate machine-readable representations of scientific findings in the form of knowledge graphs have been developed. These representations can integrate different types of experimental data from multiple papers and biological knowledge bases in a unifying data model, providing a complementary method to manual review for interacting with published knowledge. The Gene Ontology Consortium (GOC) has created a semantic modelling framework that extends individual functional gene annotations to structured descriptions of causal networks representing biological processes (Gene Ontology Causal Activity Modelling, or GO-CAM). In this study, we explored whether the GO-CAM framework could represent knowledge of the causal relationships between environmental inputs, neural circuits and behavior in the model nematode *C. elegans* (*C. elegans* Neural Circuit Causal Activity Modelling (*Ce*N-CAM)). We found that, given extensions to several relevant ontologies, a wide variety of author statements from the literature about the neural circuit basis of egg-laying and carbon dioxide (CO_2) avoidance behaviors could be faithfully represented with *Ce*N-CAM. Through this process, we were able to generate generic data models for several categories of experimental results. We also generated representations of multisensory integration and sensory adaptation, and we discuss how semantic modelling may be used to functionally annotate the *C. elegans* connectome. Thus, Gene Ontology-based semantic modelling has the potential to support various machine-readable representations of neurobiological knowledge.

Keywords:

Knowledge Graph, Semantic Modelling, Neurobiology, *Caenorhabditis elegans*, Gene Ontology

Graphical Abstract

Introduction

Caenorhabditis elegans as a Model for Systems Neuroscience

A major goal of modern neuroscience is to explain the relationship between environmental inputs and

complex behaviors in terms of the properties of their underlying neural systems. *C. elegans* has been a

productive model for neuroscience due to its wide range of easily measured behaviors, genetic tractability,

and highly stereotyped anatomy. The function of individual *C. elegans* neurons has been studied by a variety

of methods, including selective neuron ablation, either with laser microbeam irradiation (Chalfie *et al*. 1985;

Bargmann and Horvitz 1991; Bargmann and Avery 1995; Liu and Sternberg 1995) or genetically encoded

cell killing (Harbinder et al. 1997; Srinivasan et al. 2012). These physical studies, complemented by genetic

screens resulting in mutant animals with distinct behavioral and neuronal phenotypes, have implicated

specific neurons in behaviors (Bargmann 1993) and identified genes and neurons required for responses to

environmental or pharmacological inputs (Waggoner *et al*. 1998). Technological advances, such as cell-

specific application of optogenetic and chemical perturbations (Husson *et al*. 2013; Pokala *et al*. 2014) in

combination with calcium imaging of individual neurons (Chung *et al*. 2013), have begun to outline the

causal relationships between neurons, both locally and via long-range connections (Shen *et al*. 2016), while

calcium imaging allows the effect of physical inputs on neural activity to be determined. Thus, causal

relationships can be traced from inputs through neural circuits to behavior. In addition, traditional molecular

genetic methods enable the biochemical basis of these causal relationships to be elucidated. Understanding

molecular participants is particularly important for the functional description of extra-synaptic connections

because they cannot be described by anatomy or gene expression alone, yet they exert powerful effects on

neuronal activity (Bargmann 2012; Marder 2012; Bentley *et al*. 2016). In combination, the physical and

molecular data allow detailed description of *C. elegans* neural circuits underlying particular behaviors.

The GO-CAM framework can be used to Represent Causal Relationships in Biology

Given the volume of biological knowledge, a method to integrate diverse types of data into causal models of

biological systems, expressed in a common, machine-readable language, is highly desirable. A promising

method suitable for this application has been developed. The Gene Ontology (GO) Consortium has created a

semantic modelling framework for annotating causal relationships between molecular activities in the

context of functional gene annotation, known as GO-CAM (Gene Ontology Causal Activity Modelling)

(Thomas *et al.* 2019).

Semantic models (also known as knowledge graphs) are machine-readable representations of knowledge in a

given field, in which the edges of the graph describe the logical relationships between entities that comprise

a field of study. In GO-CAM, curated knowledge of gene functions annotated using the Gene Ontology and

other biologically relevant ontologies are used to create activity flow models of biological systems (Fig. 1)

(Le Novère et al. 2009). In these graphs, the logical relationships are described via a formalism known as a

semantic triple (subject-predicate-object[1]). These models can be thought of as compositions of assertions in

the form of semantic triples. For instance, the assertion "[*G-protein coupled receptor activity* (GO:0004930)]

has input [*2-heptanone* (CHEBI: 5672)]" is a semantic triple that could be included in a GO-CAM. The

semantic triple format allows edges to connect many different kinds of entities, including anatomy terms and

biological processes. For instance, "[*glucose-6-phosphate isomerase activity* (GO:0004374)] *part of*

[*canonical glycolysis* (GO:0061621)] *occurs in* [*cytosol* (GO:0005829)]" is a pair of semantic triples that

connects a GO molecular function to both a higher-level biological process and an anatomical compartment.

The Gene Ontology itself follows a hierarchical structure described with semantic triples, e.g. "[*G-protein

coupled receptor activity* (GO:0004930)] *is a* [*transmembrane signalling receptor* (GO:0004888)]" (here the

relation '*is a*' describes a child-parent relationship). This formalism allows different kinds of entities to be

connected to one another in a machine-readable format, allowing combinatorial queries and other

computational analyses.

In GO-CAM, curated knowledge of gene functions annotated using the Gene Ontology and other

biologically relevant ontologies are used to create knowledge graphs of biological systems (Fig. 1). This

framework extends traditional gene function annotation by capturing the causal flow of molecular activities,

e.g. protein kinase activity or ion channel activity, using causal relations from the Relations Ontology (RO)

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[1] Predicate and relation have the same meaning (predicate is the formal term for describing a triple).

and representing these interactions in the context of the relevant biological process and anatomy (Smith *et al*.

2005) (Box 1). These causal networks allow more in-depth computational analyses of a system than a set of

stand-alone associations between genes and ontology terms, and have the potential to bridge the gap between

biochemical and anatomical networks. Here, we explored whether the causal GO-CAM framework can

enable the representation of the causal relationships between environmental inputs, neural circuits and

behavior at varying levels of detail.

*Ce*N-CAM: GO-CAM Representation of *C. elegans* Neurobiological Knowledge

As for standard GO annotations, assertions in a GO-CAM are supported by evidence statements, ideally

experimental evidence from the published literature (Ashburner *et al*. 2000; Giglio *et al*. 2019; The Gene

Ontology Consortium 2021). To adapt the GO-CAM framework for modelling neurobiological statements

about *C. elegans* egg-laying and carbon dioxide (CO_2)-sensing behaviors, we selected a subset of relevant

papers from the *C. elegans* bibliography and identified author statements that could be used to support

construction of semantically rigorous, causal models. For the egg-laying circuit, these statements largely

involve interactions among interneurons, motor neurons, and the egg-laying apparatus, e.g. vulval muscles

and epithelia. The CO_2 avoidance circuit is focused on sensory neurons, their interaction with the

environment, and subsequent effects on locomotory behavior.

Methods & Materials

To model neurobiological processes, we began by collecting author statements from published references. In

order to ensure that our findings were broadly applicable, we collected statements from the literature on two

circuits, one centred on interneurons and motor neurons (egg-laying) and one centred on sensory neurons

(CO_2 avoidance). For the egg-laying circuit we compiled 20 papers, and for CO_2 avoidance, 8 papers. We

chose statements manually, according to a few criteria. To begin with, we chose statements that provided a

clear interpretation and that we therefore expected to be straightforward to model with GO-CAM. Later, we

selected statements describing phenomena (e.g. multi-sensory integration, neuromodulation) that were

missing from the initial dataset.

We defined an author statement as text describing: i) either an experiment or hypotheses, ii) an experimental

observation or result, and iii) a clear biological interpretation of the result. These typically comprised a

paragraph. We then attempted to model the interpretation, along with supporting evidence using the

Evidence and Conclusion Ontology (ECO) (Giglio *et al.* 2019) wherever possible. We avoided modelling

speculative suggestions that went beyond the supporting evidence.

For each author statement, we attempted to generate one or more simple assertions (i.e. semantic triples or

subject-predicate-object) that accurately modelled the author statement using classes from biological

ontologies (Table 1) including the GO (Ashburner *et al.* 2000; The Gene Ontology Consortium 2021), the

Chemicals of Biological Interest ontology (ChEBI) (Hastings et al. 2016), the Environmental Conditions,

Treatments & Exposures Ontology (ECTO) (Chan *et al*. 2022), and the *C. elegans* Cell and Anatomy

Ontology (WBbt) (Lee and Sternberg 2003). In a semantic triple, these classes are connected by relations

from the Relations Ontology (Smith *et al.* 2005) (Box 1).

We collected author statements and their corresponding semantic triples into a dataframe such that the triple

representation can be read from left to right (unless otherwise specified). Supplementary Tables 1 and 2

provide the full list of author statements that were modelled for the egg-laying (91 unique statements

comprising 128 entries from 20 papers) and CO_2-avoidance circuits (59 unique statements comprising 99

entries from 8 papers), respectively. Table 2 enumerates detailed categories of biological phenomena

captured by this approach. We used this categorization process to determine whether existing ontologies

contained a sufficiently rich set of classes and whether existing RO terms were adequate to describe the

relations between classes. Where applicable, we generated definitions for required novel classes and their

necessary parents (Table 3). We then created illustrations of several useful examples.

In generating our empirical models, we sought as far as possible to ensure that all relations followed the

conventions of the GO-CAM data model. Namely, two GO Molecular Functions can be linked by causal

relations, whereas a GO Molecular Function (MF) and a GO Biological Process (BP) are linked by

mereological relations (e.g. *part of*). In addition, two BPs can be linked by mereological relations when one

BPs is part of another BP (i.e. a subprocess of the other). We also found it necessary, in some cases, to link

distinct BPs using causal relations to accurately describe the complexity of the biology. For instance, one

neuron activating another via optogenetics can be modelled by a *membrane depolarization* process causally

upstream of another *membrane depolarization* process (e.g Fig. 4A). We sought to include whichever MFs

or BPs were implied by an author statement, even if the gene was missing, or the BP was not explicitly

discussed, in order to denote missing information. We chose the most specific relation or GO term that we

felt was justified in the circumstances. For instance, when modelling individual author statements, we used

causally upstream of, but when modelling compilations of statements from separate papers, we were able to

use the child term *positively regulates*. In generating our generic template models, we chose the highest level

relations and GO terms that could reasonably represent a given statement category.

Results

***Ce*N-CAM: GO-CAM provides a framework to model neurobiological statements**

As a first step in converting information from the scientific literature to a causal model using the GO-CAM

framework, we created semantic triples to represent author statements (Supplementary Tables 1 and 2). As

an example, a statement by Banerjee *et al.* describing the results of an optogenetic experiment that activates

membrane depolarization in uv1 neurons shows that the uv1 cells control the duration of egg deposition

during egg-laying behavior. We created a semantic triples to represent this finding: [*membrane*

depolarization (GO:0051899)] *occurs in* [*uv1* (WBbt:0006791)] *part of* [*negative regulation of egg*

deposition (GO: proposed)] *part of* [*egg-laying behavior* (GO:0018991)] (Supplementary Table 1, local

identifier EL12). In creating triples for 123 egg-laying and 98 CO_2 avoidance author statements, we found

that the set of relations used in the GO-CAM data model were sufficient to model all author statements in

our dataset. However, we required new classes in several other ontologies (the Gene Ontology (GO), the

Evidence & Conclusion Ontology (ECO), and the Environmental Conditions, Treatments & Exposures

Ontology (ECTO) (Chan *et al*. 2022)) to describe some statements in both datasets (25/123 statements in the

egg-laying dataset, and 84/99 in CO_2 avoidance) (Table 3). These results show that author statements

185 describing *C. elegans* neurobiology can be faithfully captured using the framework of the GO-CAM data

186 model.

188 We also found it necessary to re-evaluate some existing definitions and classifications of biological

189 processes under the GO class *behavior* (GO:0007610). For example since the primary term *oviposition* is a

190 subclass of *reproductive behavior* (GO:0019098) in GO and oviposition can be used to describe both the

191 entire behavior of egg laying and to describe the actual deposition of an egg onto a substrate, we requested to

192 switch the primary label of *oviposition* (GO:0046662) with the GO synonym *egg-laying behavior*. We also

193 requested a refinement of the definition of *egg-laying behavior* to 'A reproductive behavior that results in

194 the deposition of eggs (either fertilized or not) upon a surface or into a medium such as water'. In addition,

195 we created a new term *egg deposition* (GO:0160027), defined as 'The multicellular organismal reproductive

196 process that results in the movement of an egg from within an organism into the external environment'. In

197 this way, the mechanical process of *egg deposition* is clearly distinguished from *egg-laying behavior*, which

198 includes its regulation by the nervous system. We requested new terms for the positive and negative

199 regulation of egg deposition, defined as nervous system processes. In addition, we proposed definitions for

200 new classes required to describe CO_2 avoidance, including *carbon dioxide avoidance behavior* and its parent

201 *behavioral response to carbon dioxide* (Table 3).

203 Many statements describe findings from genetic perturbations, implicating specific pathways, whereas

204 others, such as cell ablation, leave open a variety of genetic mechanisms by which a phenotype is

205 manifested. Here, we describe the use of different relations and processes to refine models according to the

206 range of conclusions available in each case.

208 **Statement Category: Linking Neurons, Cellular and Molecular Processes, and Behaviors**

209 Fully elucidating functional neural circuits requires an understanding of the cells (e.g. neurons and muscles)

210 involved in the behavior, the molecular basis of the behavior (e.g. the relevant gene products and their

211 activities), and the coordinated relationships among them to affect the behavior. As with all biological

processes, however, the full understanding of a neural circuit and a behavior is produced from individual,

granular observations that, together and over time, combine to complete the picture. Leading up to a

complete understanding, we need to also have the ability to represent the current state of knowledge at the

organismal, anatomical, cellular and molecular level. Thus, in our first category of statements, we aimed to

capture atomized statements that link cells and genes to cellular and molecular level processes and those

processes to a specific behavior.

A traditional experiment for linking neurons to behavior is to ablate a neuron of interest and observe

behavioral effects, an experiment that gives us information at the cellular level (Chalfie *et al.* 1985). When

an ablation results in a behavioral change, it is interpreted that one or more processes (either in series or in

parallel) occurring in that cell has a causal effect on the behavior (Table 4A). Since cell ablation disrupts

unknown cellular processes, we chose to model this result using the high level GO biological process term

cellular process (GO:0009987), and the *occurs in* (BFO:0000066) relation to contextualize the cellular

process with respect to the ablated neuron. We then used the children of the broader causal relation *causally*

upstream of or within (RO:0002418*)* (or preferably a *positive* (RO:0004047) or *negative* (RO:0004046)

effect child term) to tie the *cellular process* to a *nervous system process* (GO:0050877). In the example

shown in Figure 2A (corresponding to the statement in Table 4A), this *nervous system process* corresponds

to the Biological Process term *positive regulation of egg deposition.* We used the *part of* relation in cases

where more specific perturbations were made (e.g. neuronal activation or inhibition, genetic knockouts and

rescues), allowing an assertion about the composition of the processes involved.

For an illustrative example of this distinction, it is useful to consider experiments from our collection that

generated insights by deletion and cell-specific rescue of genes involved in neurotransmitter biosynthesis.

We reasoned that since the biosynthesis can proceed even while the neuron is at rest (i.e. independent of the

induction of behavior), it should not be considered *part of* the asserted *nervous system process*, but *causally*

upstream of, positive effect (RO:0002304) (Fig. 2B, Table 4B) to a secretion process that is *part of* the

nervous system process (on the assumption that this secretion depends on the depolarization of the neuron).

We provide a more detailed explanation of how results of this type are modelled in a later section.

A more recent experimental technology for discerning the effect of neurons on behavior is optogenetic

activation. In these experiments, a specific neuron is activated by opening the light-sensitive

Channelrhodopsin ion channel, transgenically expressed in specific neurons of interest (Guo *et al*. 2009). We

modelled these results similarly to cell ablation, except that in this case, we were able to say that the

membrane depolarization that *occurs in* a specific cell is *part of* the *nervous system process* (in this case, the

negative regulation of egg deposition) that regulates *egg-laying behavior* (Fig. 2C, Table 4C)*.

Statement Category: Inputs to Neural Activity & Behavior

A second category of experiment provides insight into the molecular basis of behavior or neural activity

induced by an environmental or internal stimulus. In this type of study, a behavior or neural activity that is

typically induced by some environmental or experimental (i.e. pharmacological) condition is eliminated

under the same conditions when a gene is inactivated. The gene activity is often tied to a cell via rescue of a

behavioral mutant phenotype by cell-specific expression of the wild-type allele in the loss-of-function

background.

In these cases, we can tie the rescue gene functions to cells, (e.g. in Figure 3A, [*G protein-coupled serotonin*

receptor signaling pathway (GO:0098664)] *occurs in* [*VM* (WBbt:0006917)]), and to implied GO biological

process terms via *part of* (e.g. in Figure 3B [*intracellular receptor signalling pathway* (GO:0030522)] *part*

of [*positive regulation of negative chemotaxis* (GO:0050924)]. In contrast to the case of cell ablation, where

unknown cellular processes are disrupted, these more specific biological or cellular process terms can in turn

be assigned as *part of* the *nervous system process*. Additional ontology terms and relations can be used to

further specify processes or functions. For example, the Chemicals of Biological Interest ontology (ChEBI)

contains neurotransmitter classes (e.g. *serotonin* (CHEBI:28790))*, as well as environmental chemicals (e.g.

carbon dioxide (CHEBI:16526) which may be linked to GO receptor activities or other GO molecular

functions via *has small molecule activator* (RO:0012001) (Fig. 3A. Table 4D, Fig. 3B, Table 4E).

In some cases, the response to a stimulus is measured in a neuron without knowledge of the receptor

molecule. For instance, AFD neurons respond to removal of CO_2, but the experiment does not identify the

receptor molecule (Bretscher *et al*. 2011) (Fig 3C) (Table 4J). Because the receptor molecule is unknown, a

rescue experiment cannot localise the receptor activity to a cell, meaning that the response may depend on

receptor activity in another neuron. This is indicated by the absence of a relationship between the receptor

activity and a neuron (similarly, a gene knockout experiment that disrupts neural activity without cell-

specific rescue would tie only the membrane depolarization GO term to the neuron). These examples also

demonstrate the use of a *nervous system process* term as an intermediate between the *cellular process* terms

and the *behavior* terms. For instance, in our model of the role of *npr-1* in the carbon dioxide sensing circuit,

a CO_2 receptor activity is implied, but not tied to a gene or cell (Fig. 3B). However, the *nervous system*

process term (*positive regulation of negative chemotaxis* (GO:0050924) provides a natural point of

integration by which the receptor activity (and by implication, the cell in which it acts) can be included as

part of the same neural circuit. A representative GO-CAM model can be found here[2] .

Statement Category: Neuron-to-Neuron Functional Connectivity

An additional type of information necessary for fully modelling neural circuits and behaviors is the

functional link between neurons. We were able to model statements describing functional connectivity

between neurons. For example, an optogenetic experiment in which one neuron is depolarized by a light

stimulus and electrical currents are recorded in another neuron may show how a membrane depolarization

process occurring in the upstream neuron results in a subsequent membrane depolarization process in the

downstream neuron. To capture this relationship, we can connect two *membrane depolarization*

(GO:0051899) processes to one another with the *causally upstream of, positive effect* relation (Fig. 4A, Table

4G).

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[2] http://model.geneontology.org/64e7eefa00000614

GO also contains classes sufficient to indicate that the transmission occurs through a synapse, when this is

explicitly tested by authors. For instance, Kopchock *et al.* (2021) showed a synapse-dependent inhibitory

connection between HSN and VC, using tetanus toxin to perturb synaptic transmission. This could be

modelled using the GO term *chemical synaptic transmission* (GO:0007268) or one of its children, and the

causally upstream of, negative effect (RO:0002305) relation to describe the inhibition (Fig. 4B, Table 4H). A

similar representation would be appropriate for an experiment describing increase or loss of activity from a

recorded neuron in mutants defective for synaptic transmission via mutation of *unc-13* (encodes Munc13),

which is required for synaptic vessel exocytosis (Richmond *et al.* 1999). In contrast, mutation of *unc-31*

(encodes CAPS), which disrupts dense-core vesicle exocytosis, is required for extra-synaptic transmission

(Speese *et al.* 2007)[3]. GO does not have an explicit term for extra-synaptic signaling, or neuropeptide ligand

activity. We include an example representation for an extra-synaptic peptidergic connection between two

neurons (Supplementary Figure 2C), and provide a definition for the required new GO classes (Table 3).

Finally, we include an example that illustrates how CeN-CAM models can represent sub-cellular

phenomena involved in neuron-to-neuron functional connectivity in molecular detail (Fig. 4C) (Table J-L).

This model compiles findings from Choi *et al.* (2021), who use the connection between RIA and BAG

neurons to investigate mechanisms by which neurotransmitters are loaded into synaptic vesicles.

Generic Data Models for Statement Categories

In modeling author statements, we found it possible to construct models with varying levels of detail, e.g.

cell types, gene products, etc.. For instance, Figure 4D represents a 'minimal model' of the same statement

described in Figure 3B, representing the rescue of CO_2 avoidance by expression of the *npr-1* gene in URX.

We sought to provide a set of standards for the ideal model of a given category of experimental finding. In

our view, a satisfying model will have a structure that corresponds to the conceptual framework of the field

(here, the causal flow from inputs to circuits to behavior), and will explicitly illustrate missing knowledge.

By modelling the biology that results from different categories of experimental studies, we were able to

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[3] There is still debate in the literature as to how *unc-13* and *unc-31* may regulate distinct or common processes in synaptic and extra-synaptic transmission (for instance, see Sieburth *et al.* (2007)). In addition, tetanus toxin may disrupt dense core vesicle exocytosis as well as synaptic vesicle exocytosis, as in humans (Hoogstraaten *et al.* 2020). Our modelling here reflects the interpretations of the authors.

produce such generic data models for every category (Supplementary Figures 1, 2). In these models, the

availability of GO terms and RO relations is constrained by parentage, i.e. only the generic term in the model

or one of its children should be used. Importantly, the models are intended to be flexible, i.e. editable using

the Noctua GO-CAM modelling software (Thomas *et al.*, 2019). In particular, high-level *cellular process*

and *nervous system process* terms can be attached to as many GO molecular functions and genes as required

to represent the biology. These generic models could accommodate results from both the egg-laying and CO_2

circuits, suggesting that they may be more broadly applicable to *C. elegans* neurobiology. These models can

serve as useful starting points for researchers or biocurators to generate representations of the experimental

results, with minimal prerequisite knowledge of the underlying data model.

GO-CAM can model neural circuits

Systems neuroscience seeks to understand the causal relationships between neural circuits, the behaviors

they control, and the inputs that stimulate these circuits, in molecular detail. Having established that a wide

variety of author statements describing neurobiological knowledge can be represented in semantic triples,

and describing the required GO classes, we generated a model that captures some of the causal relationships

within a single circuit. This graph represents interactions between four of the cells that influence egg-laying

behavior, from a limited subset of statements in our collection (Fig. 5).

Though this diagram does not contain all cells, or all known connections that contribute to egg-laying, it

illustrates several useful features of using the GO-CAM framework to model this biology. For instance, the

influence of AWC in the circuit is connected to the rescue of HSN inhibition through AWC-specific

expression of *tax-4* (Fenk and de Bono 2015). This presumably involves chemical output from AWC that

depends on its electrical activity; however, the author statement does not assert this specifically. Similarly,

the serotonin synthesized in HSN is likely to be causally involved in the activation of VC, via activity-

dependent release into the synapse connecting these two neurons, but this has not been demonstrated directly

– only that exogenous serotonin can substitute for the absence of HSN, where there is evidence for *tph-1-*

dependent serotonin biosynthesis (Zhang *et al.* 2008). Finally, we used two nodes to represent serotonin,

because it allows the possibility that the HSN-VC serotonergic connection may be synaptic, while the HSN-

VM connection is extra-synaptic. Thus, *Ce*N-CAM models can represent causal flow within anatomical

networks in molecular detail, at the level of what is known, supported by statements in the published

literature, and as a result, also indicate what knowledge is missing.

In addition, we show that it is possible to use more informative relations in the context of a model that

integrates various findings from the egg-laying literature, compared to those used to model individual author

statements. In the case of representing author statements, our models were restricted to the use of

information contained in those statements. Here, in the larger *Ce*N-CAM model, we are able to use relations

that reflect an overall interpretation of the biology, such as *positively regulates* (RO:0002213) (a child of

causally upstream of, positive effect) to describe interactions between processes in different neurons.

GO-CAM can model simple circuit phenomena

Many studies of neural circuits investigate the mechanistic basis for information processing capabilities in

the brain, such as the integration of inputs from multiple sensory modalities, and changes in behavior that

depend on memory of past experience. We extended our modelling efforts to represent some of these

findings, primarily from our CO_2 avoidance behavior dataset.

<u>Context-Dependence & Multisensory Integration</u>

An important function of nervous systems in any organism is the ability to execute behavioral responses in a

context-dependent manner. This requires integrating multiple kinds of environmental information,

'computing' on that information and eliciting an appropriate response. This integration may commonly be

performed either by individual neurons responsive to multiple inputs, or by small circuits of three or more

neurons, e.g. single interneurons that integrate input from multiple sensory neurons (Ghosh *et al.* 2017).

Capturing this type of integration requires relations that imply the necessity of multiple conditions toward a

single response, sometimes referred to as AND logic.

We found a relevant example in our CO_2 avoidance dataset. In one study, *tax-2*-dependent rescue of CO_2

avoidance was found to depend on the presence of food (Bretscher *et al*. 2011) (Table 5A). We considered

whether any of the GO-CAM relations can be interpreted as conveying necessity, in particular the relation

part of. When considering processes such as those represented in a model, if one process is part of another

process, then the latter process necessarily has the former process as a part (or subprocess), meaning that in

these contexts *part of* and *has part* (BFO:0000051) are inverse relations (Smith *et al.* 2005). Figures 6A-6C

show how the necessity for AND logic might be modelled. The cell-specific rescue of CO_2 avoidance via

tax-2 expression in BAG neurons, along with the inferred CO_2 receptor activity, constitutes one 'branch' of

the model. A second 'branch' represents the involvement of food, via an inferred *signal transduction*

(GO:000716) process. These two branches converge on a proposed GO term s*ignal integration process* via

part of relations, capturing their joint necessity. We chose to include a new GO biological process for their

integration (rather than having them converge on positive regulation of CO_2 avoidance) in order to represent

that the mechanism enabling the AND logic should be asserted. This representation leaves open many

possible biological models for the mechanism by which the asserted integration might occur (for example,

one in which food and CO_2 are sensed by distinct sensory neurons, and integrated in a third interneuron),

while capturing AND logic.

<u>Co-ordination of Neural Activity and Physical Features of Behavior</u>

Our egg-laying dataset contains statements describing a mechanical feature of egg-laying regulation, namely

neural activity in the vulval muscles is co-coordinated with the phase of body bending during locomotion

(Kopchock *et al*. 2021) (Table 5B). We suggest that *part of* and a new GO Biological Process class *behavior*

co-ordination process could be used to model this (Fig. 6D). This representation captures the author's

interpretation that neural activity in VM drives oviposition conditional on features of locomotory behavior.

<u>Sensory Adaptation</u>

Some important modelling challenges arose in our attempts to model sensory adaptation. This required

representing feedback regulation, and the different qualities of input (in terms of concentration and duration)

relevant at different points in the process.

Waggoner et al. (2000) describe sensory adaptation to nicotine mediated by down-regulation of the UNC-29

receptor (Table 4G, 4H), in a *tpa-1*-dependent manner. Using this example, we found that feedback is best

modelled by creating a duplicate temporal instances of a given input, molecular activity and associated

biological process (Fig. 9). One set is used for the initiation of the gene expression changes that mediate

desensitization to acetylcholine (nicotine), and another for the altered behavioural response to the same

input. It was also necessary to model how this process depends on *tpa-1,* given that the statement doesn't

show precisely how the *tpa-1*-enabled activity is related to the *unc-29*-enabled activity. We found that this

was best achieved by invoking an instance of *protein kinase activity* (GO:0004672) enabled by *tpa-1* that is

part of a s*ignal transduction* (GO:0007165) process that is, in turn *part of* an instance of *sensory adaptation*

to nicotine (a proposed novel GO Biological Process), and *causally upstream of, positive effect* to [*negative*

regulation of gene expression (GO:0010269)]. We then modelled a causal effect of the *acetylcholine*

receptor signalling pathway (GO:0095500) on the *tpa-1* linked *signal transduction process* (GO:007165).

This completes the causal chain required to represent the feedback inherent in sensory adaptation, while

allowing that *tpa-1* may be required in parallel to *unc-29,* rather than downstream. We were able to capture

the requirement for sustained input for gene regulation, and the requirement for elevated input concentration

in the post-adaptation response using terms from (PATO) and the associated relation *has quality*

(RO:0000086). However, as discussed in a later section, this approach has limitations, and we would prefer

to develop a systematic set of ontology terms for temporal dimensions of environmental input hosted on

ECTO.

Neuromodulation

An important goal of neural modeling is to capture neuromodulatory effects, which may be defined as

changes in neuronal excitability or dynamics, due to changes in internal state or external context (Bargmann

2012). We found a small number of entries in our egg-laying dataset that described changes in membrane

excitability (e.g. Table 5F). We chose to model these with the GO term *regulation of resting membrane*

potential (GO:0060075), with the view that changing the ability of the cell to maintain its resting potential is

the primary mechanism for regulating neuronal excitability. However, it may be more appropriate to use a

parent GO class that can model changes in excitability, rather than implying any mechanism. For instance,

one might imagine induced changes in receptor expression that could alter excitability or responsiveness,

without changing the resting membrane potential (Shine *et al*. 2021). The term *regulation of membrane*

depolarization (GO:0003245) and its children may be more appropriate when the mechanism is not known.

Extending Existing Ontology Classes for Modelling Neurobiology

We found that the use of existing ontologies provided the correct classes for building our models of neural

circuitry. However, in some cases we found that additional classes would be useful for a complete and

accurate description of the type of biology we are modeling. These proposed additional classes would be

added to GO, the Evidence and Conclusion Ontology (ECO) and the Environmental Conditions, Treatments

and Exposures Ontology (ECTO) and are listed in Table 3. Evidence supported by four types of

experiments, chemical inhibition of neurons via histamine chloride (Pokala *et al.* 2014), inhibition of

synaptic transmission (Sweeney *et al*. 1995), mechanical perturbation, and long-term exposure experiments

require additional classes in ECO. The categories below describe the biological phenomena that require new

GO Biological Process terms, GO Molecular Function terms and ECTO terms to model. Inclusion of these

new classes would enrich the kinds of queries that could be supported by *Ce*N-CAM (for instance, we may

want a list of all interneurons whose activity is known to be modulated by peptidergic output from ASI

neuron). Particularly useful would be the addition of the previously mentioned requirement for GO terms

describing extra-synaptic neuropeptide signaling and neuropeptide activity. OBO ontologies are carefully

managed, and ontology developers provide processes for the addition of new classes. For instance, we were

able to add a GO term for *carbon dioxide receptor activity* (GO:0170015) via the GO GitHub repository by

providing   the   necessary   information   for   its   incorporation   into   the   ontology   (see

https://github.com/geneontology/go-ontology/issues/24994). We discuss other proposed classes below.

<u>Fine Temporal Dynamics of Neural Activity & Behavior</u>

Many statements described neural activity in fine temporal detail. Experimental treatments are sometimes

reported to result in changes to either magnitude, duration and/or frequency of membrane depolarization or

hyperpolarization (e.g. Table 5H). In some cases, these phenotypes lead authors to the interpretation that

these parameters of a neuron's behavior are under selection in wild-type organisms, and required to perform

the given behavioral task (for instance, changes in the frequency of calcium transients in neurons of the egg-

laying circuit are thought to reflect shifts from 'active' to 'inactive' states of the circuit, reflecting phases of

the behavior (Collins *et al.* 2016)). However, the GO class for *membrane depolarization* (GO:0051899) does

not distinguish these variations, and related terms such as *positive regulation of membrane depolarization*

(GO:1904181) explicitly groups these phenomena together under one term. In the future, it may be useful to

have these classes separated into explicit categories for a more comprehensive and informative view of how

neural activity is regulated.

Likewise, many assays of egg-laying behavior document its temporal features, dividing it into active and

inactive phases, and measuring the effect of various perturbations on their duration and frequency (e.g. Table

5I). In the CO_2 avoidance literature, a small number of entries described fine details in motor output as a

result of neuronal perturbations, such as changes in rates of reversal or frequency of omega turns (Bretscher

et al. 2011) (Table 5K). We were unable to model these features due to a lack of sufficiently fine-grained

GO terms in the Biological Process ontology. However, we note that WormBase has a phenotype ontology to

describe behavior in many of the appropriate ways (for example *turning frequency increased*

(WBPhenotype:0002313)) (Schindelman *et al*. 2011). Since these are mutant phenotypes and not Biological

Processes, these ontology classes are a poor fit for *Ce*N-CAM. Conversion of these phenotype classes into

meaningful GO Biological Processes would be helpful to create more fine-grained models of behavior.

<u>Temporal Features of Environmental Input</u>

In modelling environmental inputs, we found it necessary to model several temporal features. As discussed

previously, our sensory adaptation example required representing sustained input, and the relative decrease

in concentration required for the future behavior. We approximated these using PATO terms *chronic*

(PATO: 0001863) and *increased amount* (PATO:0000470), respectively. However, PATO lacked terms

required to model changes in input concentration or intensity over time, as required to model the OFF

response to CO_2 in ADF neurons (Fig 5B). We found that terms in the Environmental Conditions,

Treatments & Exposures Ontology (ECTO) came closer to these requirements (e.g. *exposure to decreased*

methane (ECTO:4000005)), but a specific exposure term for many chemicals, such as carbon dioxide, does

not exist. We propose and define new classes specifying temporal properties that could be hosted in ECTO

(Fig. 8) (Table 3).

Discussion

Given the size, scope and rapid growth of the biological literature, new methods are required to integrate,

represent and interpret accumulating knowledge at varying levels of detail. One method for achieving this is

to integrate objects from relevant ontologies in semantic graphs. In this work, we demonstrate the

applicability of a Gene Ontology based semantic modelling framework, GO-CAM, for representing

knowledge of neural circuits in *C. elegans*. By capturing author statements in select papers, we were able to

construct simple semantic statements and then link those statements together to begin building causal models

of two *C. elegans* behaviors, egg-laying and carbon dioxide avoidance. We found that the existing Relations

Ontology (RO) relations used in GO-CAMs are adequate, but new classes are required in several ontologies,

including the Gene Ontology (GO), the Evidence & Conclusions Ontology (ECO) and the Experimental

Conditions, Treatments & Exposures Ontology (ECTO) to fully represent the statements in our collection.

In general, the GO contains a rich vocabulary for neurobiology, in part due to projects such as SynGO

(Koopmans *et al*. 2019), which expanded GO's representation of synaptic function, and deposited

corresponding annotations in the GO repository as GO-CAM models. In addition, the Reactome

knowledgebase contains pathways for synaptic transmission, and these have been converted to GO-CAMs

(Good et al. 2021). To complement the synaptic transmission part of the ontology, new terms will be

required to describe features of extra-synaptic (i.e. peptidergic) connectivity. We also anticipate a more

widespread need to model temporal details of sensory neuron input, since chemotactic behaviors typically

involve sensing of spatial gradients, experienced by sensory neurons as change over time, resulting in

movement towards or away from the odour source. For instance, the sensory neuron AWA adapts to a given

concentration of diacetyl, requiring increasing concentration for continued depolarization and associated

positive chemotaxis (Larsch *et al.* 2015). Adding these temporal details to the inputs of individual neurons

would allow for more expressive representations. In addition, many of the GO Biological Process terms that

we propose as additions to the Gene Ontology are the result of describing the processes at the level of an

organism or cell and are not derived from attempts to annotate gene function. Such temporal details are often

derived from phenotypic measurements resulting from non-genetic perturbation (e.g. cell ablation,

pharmacological inputs), in anticipation of the involvement of gene activities in the programmed regulation

of these processes. In practice, new GO BP terms based on these observations will likely need genetic

evidence before they can be included in the GO; however, we include them here as suggestions, which may

guide future proposals as the need arises.

In addition, the models presented here go beyond the minimal requirements for the conversion of author

statements into semantic triple format. According to our criteria, a satisfying model should reflect the

conceptual framework of the field (in this case, representing causal flow from inputs through circuits to

behavior). In this way, the models indicate which knowledge is missing. For instance, Figure 3B depicts the

role of *npr-1* in the URX neuron in carbon dioxide avoidance behavior. By including a *nervous system*

process term indicating the involvement of neural circuit, it is possible to indicate that a carbon dioxide

receptor, whose encoding gene and cellular site of action require identification, are part of the circuit. The

data modelling work presented here also provided us with an empirical basis for creating generic models or

templates for each of the statement categories described above (Supplementary Figs. 1, 2). In constructing

these generic models, we followed structures that reflected the relevant conceptual framework into which

particular classes of experimental results should fit. For instance, the full description of a peptidergic

connection between neurons should involve the relevant ligand(s), receptor(s), ion channel(s) and encoding

genes (Supplementary Fig. 2C). Including the overarching biological process term *neuron-to-neuron*

signaling by neuropeptide allows a database to be indexed for these types of connections. In this way,

scientists and biocurators can collaborate to generate models with a common understanding of their proper

criteria.

We also tried to capture simple 'computations' important for nervous system function, and arrived at some

modelling principles that are noteworthy. Firstly, when representing the AND logic involved in multisensory

integration, it is important to use relations that convey necessity, and have separate causal flows that

converge on a single biological process. We note that the proposed GO Biological Process terms (*signal*

integration process and *behavior coordination process*) describe an information processing event that could

in theory be carried out via any molecular mechanism that satisfies the task. Representing similar kinds of

neurobiological knowledge in the GO may require further understanding of the types of molecular

mechanisms that typically underlie this type of nervous system process (Ghosh *et al.* 2017). We also found

that, when representing feedback (as in our sensory adaptation example), duplicating the representation of

molecular functions and/or biological processes, occurring in defined temporal sequences, is important for

clear and consistent modelling.

In this study, we focused on modelling interactions within neural circuits, and their relationship to broad

features of behavior, rather than the detailed mechanics of motor programs that they control. In principle, it

is possible to link neural activities to the mechanical outputs of neural activity, where both are considered

part of the organismal behavior under study. In the case of egg-laying, this motor output is simple, involving

only the contraction of the vulval muscles. However, CO_2 avoidance involves a complex series of

locomotory processes, each of which is regulated by specific patterns of neural activity (for example, see

Bretscher *et al.* 2011). As discussed above, inference of new biological process terms by conversion of the

appropriate terms from the *C. elegans* Phenotype Ontology will allow modelling of these features of

behavior. These motor outputs could then be modelled as *part of* the organismal behavior *carbon dioxide*

avoidance behavior (i.e. they are the targets of the *regulation of chemotaxis* term in the models diagrammed

here).

One limitation not previously discussed is that GO-CAM currently has no way of incorporating negative

data. In some cases, this prevented documentation of important discoveries from our literature search. For

instance, (Shyn *et al*. 2003) found that in the absence of VC neurons and HSN neurons, spontaneous Ca^{2+}

transients continued in the vulval muscles, suggesting that these neurons are not necessary for VM activity

(Table 5J). These are arguably important omissions from these knowledge graphs.

With these adjustments, this work demonstrates the possibility of creating a machine-readable knowledge

base for neurobiology that can return information based on queries. An important part of this resource will be

to generate a representation of the *C. elegans* brain that is computable, since the current anatomy ontology

does not contain synaptic or gap junction connections between neurons (Lee and Sternberg 2003).

Incorporating connectome data that contains the appropriate neuron to neuron relations and property chain

algebra (i.e. (Neuron A synapses to Neuron B) and (Neuron B synapses to Neuron C) implies that (Neuron A

connects with Neuron C)) will allow queries that include or depend on synaptic connectivity information.

The application and widespread use of this technology depends on the amount of information incorporated

into the knowledgebase, much of which at this point is directly dependent on manual input by curators.

Given our definition of an author statement as a passage of text following a stereotyped form (hypothesis,

observation, interpretation), it is possible to envision how author statements could be identified

automatically. We envision a scenario in which machine intelligence could be applied to identify not only

author statements, but identify the category of experiment they describe, and the GO terms that correspond to

words within them. Using the generic data models described here as templates could help to ensure that

machine-generated models are constrained by a desirable structure. With these capabilities, a large volume of

the *C. elegans* neural circuit literature could potentially be converted into *Ce*N-CAM models

computationally. The author statements that we collected as part of this work will serve as training data to

pursue this type of approach.

It is also important for biologists to have usable and intuitive ways of interacting with and analysing

synthesized knowledge. One way to achieve this is by representing compiled neurobiological data in an

anatomical context. For instance, the Virtual Fly Brain project has used an ontology-based approach to

integrate connectivity and single-cell gene expression data, which can be visualised in a 3-dimensional

visualization of the brain, using a semantic integration framework (Milyaev et al. 2012; Court et al. 2023).

This allows users to run queries to explore gene expression and phenotype data in an anatomic context. We

are exploring the possibility of functionally annotating the *C. elegans* connectome in molecular detail using

*Ce*N-CAM (Fig. 9). The relevant data are the same as those captured by the statement categories for which

we have generated templates, namely causal relationships between inputs to neurons, neurons to behavior,

and causal connection between neurons. In addition to populating template data models, the GO terms in the

relevant author statements could be used to populate a dataframe of the kind used by visualization software

such as Cytoscape (Shannon *et al*. 2003)(Supplementary Figure 3), ideally in an automated manner. This

visualization could serve as an intuitive entry point for exploring neural circuit function on a connectome

scale, where evidence behind individual elements of the graph could be accessed by linking to the

corresponding *Ce*N-CAM models. An anatomical visualization that includes functional and connectivity data

would allow predictions to be made about functional relationships between different circuits. For instance,

CO_2 has been shown to inhibit egg-laying (Fenk and de Bono 2015) in an AWC-dependent manner.

Representing neurons that respond to CO_2 along with neurons that control egg-laying in a connectome

context (Fig. 9) suggests that ASH is a CO_2 responsive neuron synaptically linked to HSN. Indeed, ASH was

later shown to inhibit both egg-laying and HSN activity (Wen et al. 2020). Functional connectome

annotation would also enable various kinds of system-wide analysis of the *C. elegans* brain – a research

avenue that has so far been pursued in the absence of functional information (Reigl *et al*. 2004; Alon 2007;

Jarrell *et al*. 2012). We also envision the ability to make useful predictions using the underlying semantic

models. For instance, the graphs may include causal links between molecular functions and behaviors that

result from synthesis of disparate literature, leading to new predictions about how genetic or

pharmacological perturbations may affect behavior. Thus, the work described here provides semantically

and biologically rigorous foundations for an integrated systems neuroscience resource combining knowledge

representation, connectome annotation and associated computational analyses of *C. elegans* nervous system

function.

Declarations

Abbreviations

GO (Gene Ontology), GOC (Gene Ontology Consortium), GO-CAM (Gene Ontology Causal Activity

Modeling), CeN-CAM (*Caenorhabditis elegans* Neural Circuit Causal Activity Modeling), ChEBI

(Chemicals of Biological Interest Ontology), Relations Ontology (RO), Evidence & Conclusions Ontology

(ECO), Basic Formal Ontology (BFO), Phenotype and Trait Ontology (PATO), Environmental Conditions,

Treatments and Exposure Ontology (ECTO).

Data & Materials Availability

All data generated or analysed during this study are included in this published article (and its supplementary

information files).

Competing Interests

The authors declare that they have no competing interests

Author Contributions

P.W.S & S.J.P. conceived and designed the study, S.J.P, K.V.A & D.P.H. conducted the study. S.J.P, K.V.A,

D.P.H & P.W.S wrote and revised the manuscript. S.J.P. prepared all the figures. All authors approved the

manuscript.

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Tables

Table 1: Biological Ontologies Used To Generate CeN-CAM models

GO-CAM element	Ontology	Example
Molecular activity	GO molecular function	serotonin receptor activity (GO:0099589)
Biological process	GO biological process	Membrane depolarization (GO:0051899)
Location	GO cellular component	cytosol (GO:0005829)
Cell	WormBase Anatomy Ontology	HSN (WBbt:0006830)

Active Entity (Gene/Gene Product)	WormBase	*tph-1* (WBGene00006600)
Chemical inputs	Chemical Entities of Biological Interest (ChEBI)	dioxygen (CHEBI:15379)
Relations arrows	Relations Ontology (RO), Basic Formal Ontology (BFO)	occurs in (RO_0002479)
Evidence codes	Evidence & Conclusions Ontology (ECO)	optogenetic evidence used in manual assertion (ECO:0006033)
Environmental Conditions	Phenotype and Trait Ontology (PATO) Environmental Conditions, Treatments and Exposure Ontology (ECTO)	Increased duration (PATO_0000498)

771
772   **Table 2: Categories of Neurobiological Phenomena Modelled with GO-CAM**
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Neuronal basis of behavior	Receptor or G-protein activity regulates behavior or cell activity
Neuron regulates behavior	G protein activity in specific neuron regulates behavior
Cellular process regulates behavior	GPCR regulates G -protein-activity in specific neuron to regulate behavior
Neuronal activity regulates behavior	Neuromodulation of specific neuron by G protein signaling
Neuronal activity dependent secretion from identified neuron regulates behavior	G protein activity regulates gene expression
Neuronal activity dependent neurotransmitter secretion	G protein activity regulates gene expression cell autonomously
Neuron-neuron interaction	G protein activity regulates neurotransmitter biosynthesis
Activity of Neuron A regulates activity of Neuron B	G protein activity regulates phospholipase activation

Activity of Neuron A regulates activity of Neuron B (synapse-dependent)	G protein signaling activity regulates neuronal activity cell autonomously
Mechanical stimulation of Neuron A regulates activity of Neuron B	GPCR regulates ion channel
Negative autoregulation of neuronal activity	Ion channel regulates neuronal activity via GPCR
Neural activity depends on extra-synaptic signaling	G protein activity regulates neurotransmitter biosynthesis cell autonomously
Environmental influence on behavior or cell activity	GPCR regulates G -protein-activity
Environmental input regulates behavior	G protein activity regulates neurotransmitter biosynthesis
Environmental input regulates neuronal activity	Neuromodulation of specific neuron by G protein signaling
Environmental condition regulates neuronal activity	Receptor activity regulates neuronal activity cell autonomously
Environmental condition regulates behavior	**Cellular Process**
Mechanical process regulates neural activity	Neurotransmitter biosynthesis
Environmental input regulates behavior via defined neuron	Neurotransmitter signaling pathway affects behavior
Environmental input regulates gene expression	Biochemical process regulates neural activity
Receptor-ligand Interaction	Dense core vesicle exocytosis from identified neuron regulates behavior
Receptor-ligand interaction	Gene activity regulates neural activity
Neurotransmitter regulates neuronal activity via ion channel	Gene activity in identified neuron regulates behavior
Neurotransmitter regulates behavior via specific receptor	Dense core vesicle exocytosis regulates behavior
Neurotransmitter regulates behavior via ion channel in identified neuron	Neuropeptide signaling pathway affects behavior
Neurotransmitter regulates behavior via specific receptor in identified neuron	Neuropeptide signaling pathway affects behavior via identified cell
Neurotransmitter affects identified receptor class	Regulation of gene expression in identified neuron
Ion channel activity regulates behavior or cell activity	**Neurotransmitter/neuropeptide activity regulates behavior or cell activity**
Ion channel regulates neural activity	Neurotransmitter biosynthesis from identified source neuron regulates behavior
Neuromodulation of specific neuron by ion channel activity	Neurotransmitter biosynthesis regulates behavior
Ion channel regulates membrane potential	Neuropeptide from specific neuron regulates behavior
Ion channel activity in defined neuron regulates behavior	Neurotransmitter activity depends on ion channel
Ion channel activity regulates behavior	Neurotransmitter regulates behavior
Nervous System Process	Neurotransmitter regulates neuronal activity
Adaptation to chemical stimulus	Regulation of neurotransmitter activity by upstream neuropeptide activity
Co-ordination of locomotion and neural activity to influence behavior	Regulation of secretion by upstream neuropeptide activity

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810 **Table 3: Definitions & Classification for Proposed New GO Classes**
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Category	Proposed Modified Classes	Modification	Classification
GO Biological Process	*Oviposition (GO:0018991)*	Change primary term name to synonym 'egg-laying behavior' New definition: The muscle system process resulting in the deposition of eggs (either fertilized or not) upon a surface or into a medium such as water.	is a *muscle system process (GO:0003012)*, part of 'egg-laying behavior'
GO Biological Process	*Behavior (GO:0007610)*	New classification (see right)	is a *response to stimulus (GO:0050896)*

Category	Proposed New Class	Definitions	Classification
GO Biological Process	Egg deposition	The multicellular organismal reproductive process that results in the movement of an egg from within an organism into the external environment.	is a *reproductive behavior (GO:00198098); part of oviposition/egg-laying behavior (GO:0018991)*

GO Biological Process	Positive regulation of egg deposition	Any process that positively regulates the rate, frequency or extent of egg deposition	is a *nervous system process (GO:0050877); part of oviposition/egg-laying behavior (GO:0018991)*
GO Biological Process	Negative regulation of egg deposition	Any process that negatively regulates the rate, frequency or extent of egg deposition	is a *nervous system process (GO:0050877); part of oviposition/egg-laying behavior (GO:0018991)*
GO Biological Process	Neuron-to-neuron extra-synaptic peptide signaling	Any process by which a cellular process within one neuron influences a cellular process within another neuron via a secreted gene product, where this secretion occurs independently of synapses	is a *neuron-to-neuron chemical signaling process*
GO Biological Process	Neuron-to-neuron chemical signaling	Any process by which a cellular process within one neuron influences a cellular process within another neuron via a secreted molecule	is a *cell-cell signaling (GO:0007267) process*
GO Biological Process	Behavioral response to carbon dioxide	The behavior of an organism in response to a carbon dioxide stimulus.	is a *behavior (GO:007610)*; is a *response to carbon dioxide (GO:0010037)*
GO Biological Process	Carbon dioxide avoidance behavior	The behavioral response to carbon dioxide which results in the directed movement of a motile cell or organism towards a lower carbon dioxide concentration	is a *negative chemotaxis (GO:0050919)*; is a *behavioral response to carbon dioxide*
GO Biological Process	Negative regulation of carbon dioxide avoidance behavior	Any process that negatively influences locomotory behavior directed away from a source or gradient of carbon dioxide .	is a *nervous system process (GO:0050877); part of behavioral response to carbon dioxide*
GO Biological Process	Positive regulation of carbon dioxide avoidance behavior	Any process that positively influences locomotory behavior directed away from a source or gradient of carbon dioxide.	is a *nervous system process (GO:0050877); part of behavioral response to carbon dioxide*
GO Biological Process	Adaptation of neuron to stimulus	Any process that results in an increased threshold for induction of neural activity due to prior exposure to the same stimulus	is a *negative regulation of membrane depolarization (GO1904180)*
GO Biological Process	Sensitization of neuronal response to stimulus	Any process that results in a reduced threshold for induction of neural activity due to prior exposure to the same stimulus.	is a *positive regulation of response to stimulus (GO:0048584)*
GO Biological Process	Sensitization of behavioral response to stimulus	Any process that results in a reduced threshold for induction of behavioral response due to prior exposure to the same stimulus.	is a *positive regulation of behavior (GO:0048520)*
GO Biological Process	Sensory adaptation in behavioral response to stimulus	Any process that results in an increased threshold for induction of behavioral response due to prior exposure to the same stimulus.	is a *negative regulation of behavior (GO:0048521)*
GO Biological Process	Behavior co-ordination process	Any neural process that links the execution or cessation of one behavior to the induction or cessation of another behavior.	is a *regulation of behavior (GO:0050795)*

GO Biological Process	Signal integration Process	Any nervous system process by which different types of input to the nervous system contribute in combination to a behavioral or physiological output.	is a *nervous system process (GO:0050877)*
GO Molecular Function	CO2 receptor activity	Binding to and responding, e.g. by conformational change, to changes in the cellular level of carbon dioxide (CO2) or its dissociation products in water.	is a *signaling receptor activity (GO:0038023)*
ECO Evidence Class	Neuron Chemical Inhibition Assay Evidence used in Manual Assertion	A type of experimental phenotypic evidence that is used in a manual assertion, arising from experiments in which the output from a neuron is inhibited by a chemical	is a *experimental phenotypic evidence (ECO:0000059)*
ECO Evidence Class	Synaptic Transmission Inhibition Evidence used in Manual Assertion	A type of experimental phenotypic evidence, that is used in a manual assertion, arising from experiment in which neuron-to-neuron synaptic transmission is manipulated using inhibitors of synaptic transmission, that is used in a manual assertion.	is a *experimental phenotypic evidence (ECO:0000059)*
ECO Evidence Class	Mechanical Perturbation Evidence used in Manual Assertion	A type of experimental phenotypic evidence, that is used in a manual assertion, arising from experiment in which cellular responses are manipulated using mechanical force, , that is used in a manual assertion.	is a *experimental phenotypic evidence (ECO:0000059)*
ECO Evidence Class	Long-term Exposure or Conditioning Evidence used in Manual Assertion	A type of experimental phenotypic evidence, that is used in a manual assertion, arising from experimental treatment involving sustained exposure of an organism to one or more environmental conditions.	is a *experimental phenotypic evidence (ECO:0000059)*
ECTO Class	exposure to increasing carbon dioxide	An exposure event involving the interaction of an exposure receptor to increasing amount of carbon dioxide	is a *exposure to chemical (ECTO: 0000231)*
ECTO Class	exposure to decreasing carbon dioxide	An exposure event involving the interaction of an exposure receptor to increasing amount of carbon dioxide	is a *exposure to chemical (ECTO: 0000231)*

813   **Table 4: Author Statements Collection A**

	Author	Author Statement
A	Waggoner *et al.* 1998	"The roles of individual neurons in controlling the timing of egg-laying events can be determined with high precision by eliminating specific neurons by laser ablation and assaying the effect of the ablation on behavior. We therefore eliminated the neurons with prominent synaptic input to the egg-laying muscles to determine how their absence affected the timing of egg-laying events. We first investigated the involvement of the HSNs, a pair of serotonergic motor neurons that are required for efficient egg laying. By tracking the behavior of animals lacking both HSNs, we found that elimination of the HSNs did not qualitatively alter the pattern of egg laying: eggs were still laid in clusters, and the intervals between clusters and between egg-laying events within a cluster were still exponentially distributed. However, HSN ablation did cause a substantial lengthening of the inactive phase, which led to a slower overall rate of egg laying (Figure 2A). Since loss of the HSNs decreased the frequency of egg-laying clusters (i.e., $\lambda 2$ was decreased; Table 1) but did not slow the egg-laying rate within these clusters ($\lambda 1$ was actually increased), these results suggest that the HSNs stimulate egg laying by inducing the active state."
B	Bany *et al.* 2003	"Because the VC neurons appear to inhibit egg laying and are cholinergic, we tested whether the VCs release acetylcholine to inhibit egg laying. The VCs are the only cells of the egg-laying system that express the UNC-4 complex, CHA-1, and UNC-17 (Lickteig *et al.*, 2001); however, because unc-4, cha-1, and unc-17 are each expressed in other neurons, it was necessary to determine whether mutations in these genes cause hyperactive egg laying specifically attributable to their effects on the VC neurons. For this purpose, we expressed the unc-4, cha-1,or unc-17 cDNAs in the VC neurons and determined whether this rescued the hyperactive egg-laying defects of the corresponding mutants. To direct VC expression, we used a modified lin-11 promoter similar to that used to express GFP in Figure 3A (see Materials and Methods). Expression of the unc-4 cDNA using this promoter rescued the hyperactive egg-laying defect of unc-4 mutants, returning the percentage of early-stage eggs laid to near-wild-type levels (Fig. 4A). Furthermore, expressing the cha-1 cDNA in the VC neurons of cha-1 mutants also rescued their hyperactive egg-laying phenotype (Fig. 4B). Similar experiments with unc-17 gave analogous results (data not shown). Restoring the inhibition of egg laying by restoring the ability of the VC neurons to signal with acetylcholine provides our most compelling evidence that it is the VC neurons that inhibit egg laying."
C	Banerjee *et al.* 2017	"We next sought to determine whether uv1 activation is sufficient to inhibit egg-laying. To address this question, we expressed channelrhodopsin (ChR2) in uv1 cells using the regulatory regions of ocr-2 as above. Light stimulation immediately following the initial egg-laying event of an active phase (see Methods) significantly delays subsequent egg-laying events, and also significantly reduces the total number of egg-laying events within an active phase (Fig 2)(S1 Movie). For example, under control conditions a majority (~80%) of animals show a delay between the first and second egg-laying events within an active phase of <20 s (light stimulation, -ATR) (Fig 2B). This proportion is reduced dramatically (to around 10%) when uv1 cells are activated (light stimulation, +ATR)...Taken together, our findings provide evidence that uv1-mediated inhibition of egg-laying promotes periods of quiescence in the egg-laying program and plays a key role in setting their duration."
D	Carnell *et al.* 2005	"The expression of gfp in the vulval muscles suggests that ser-1 may be acting in vulval muscles to mediate the stimulatory effect of 5-HT on egg laying. To test this hypothesis, we expressed the ser-1 cDNA using the muscle-specific myo-3 promoter (Okkema *et al.*, 1993) to determine whether it could rescue 5-HT-induced egg laying. Consistent with this hypothesis, we found the Pmyo-3::ser-1(+) transgene partially restored 5-HT-dependent egg laying to ser-1(ok345) animals (Fig. 2A). A wild-type ser-1 transgene with the same 3.4 kB promoter that failed to express gfp in the vulval muscles also failed to rescue the egg-laying defects of the ser-1 mutant animals (Fig. 2A). These results indicate that ser-1 expression in muscle can restore egg laying. Previous studies have indicated that 5-HT acts on vulval muscle to stimulate egg laying (Trent *et al.*, 1983; Brundage *et al.*, 1996; Waggoner *et al.*, 1998; Bastiani *et al.*, 2003; Shyn *et al.*, 2003). Our results indicate that ser-1 mediates this response."
E	Carillo *et al.* 2013	"NPR-1 is not expressed in BAG neurons but is expressed in a number of other sensory neurons as well as some interneurons (Macosko *et al.*, 2009). To identify the site of action for the regulation of CO2 response by npr-1, we introduced the N2 allele of npr-1 into npr-1(lf) mutants in different subsets of neurons and assayed CO2 response. We found that expressing npr-1 in neuronal subsets that included the O2-sensing URX neurons (Cheung *et al.*, 2004; Gray *et al.*, 2004) restored CO2 response (Fig. 3A). These results suggest that NPR-1 activity in URX neurons is sufficient to enable CO2 avoidance. However, we cannot exclude the possibility that NPR-1 function in other neurons also contributes to CO2 avoidance."
F	Bretscher *et al.* 2011	"Strikingly, AFD also responded to removal of CO2 with a fast Ca2+ spike that peaked within 10 s ("CO2-OFF" response...CO2-evoked activity in AFD could be due to synaptic input to AFD. To test this, we imaged CO2 responses in unc-13 mutants, which have severe defects in synaptic release (Richmond *et al.*, 1999). The AFD CO2 responses of unc-13 animals were indistinguishable from wild-type (Figures 2H and S1C). These data suggest that, as well as being a thermosensory neuron (Mori and Ohshima, 1995; Kimura *et al.*, 2004; Clark *et al.*, 2007), AFD is a CO2 sensor with both ON and OFF responses."
G	Collins *et al.* 2016	"To directly test how neurotransmitter signaling from the HSNs regulates egg-laying circuit activity, we used the egl-6 promoter to express Channelrhodopsin-2 in the HSNs (Emtage *et al.*, 2012), allowing us to drive neurotransmitter release specifically from the HSNs with blue light. ...We found that activation of HSNs resulted in circuit activity reminiscent of a spontaneous active state, including rhythmic Ca2+ activity of both VCs and vulval muscles, and egg-laying events that accompanied a subset of these Ca2+ transients... These results suggest that the high level of HSN activity after optogenetic

		activation induces strong coupling of VC and vulval muscle excitation."
H	Kopchock *et al.* 2021	"To determine whether VC synaptic transmission regulates egg laying via HSN, we recorded HSN Ca2+ activity in WT and transgenic animals expressing TeTx in the VCs (Fig. 6A). During the egg-laying active state, the HSNs drive egg laying during periods of increased Ca2+ transient frequency in the form of burst firing (Fig. 6B) (Collins *et al.*, 2016; Ravi *et al.*, 2018a). We observed a significant increase in HSN Ca2+ transient frequency when VC synaptic transmission was blocked compared with nontransgenic control animals (Fig. 6C). WT animals spent ∼11% of their time exhibiting high-frequency burst activity in the HSN neurons, whereas transgenic animals expressing TeTx in the VC neurons spent ∼21% of their time exhibiting HSN burst firing activity (Fig. 6D). These results are consistent with the interpretation that VC neurotransmission is inhibitory toward the HSNs, such as proposed in previous studies (Bany *et al.*, 2003; Zhang *et al.*, 2008)."
J	Choi *et al.* 2021	"VGLUTs are members of a family of anion transporters that move diverse solutes, including inorganic phosphate, acidic sugars, negatively charged amino acids, and phosphorylated adenosine nucleotides33. As a member of the SLC17 family of transporters, VST-1 is likely an anion transporter and there are different ways an anion transporter in the synaptic vesicle membrane could limit glutamate uptake. ...we used synaptopHluorin to measure vesicular pH in wild-type and vst-1 BAG neurons. Measurements of total and surface-accessible pHluorin (Fig. 3g) allow computation of vesicular pH42...Importantly, we found that loss of VST-1 caused a measurable increase in vesicular pH (Fig. 3h), consistent with a model in which VST-1 supports anion influx into synaptic vesicles. We also measured vesicular pH in BAG neurons lacking EAT-4/VGLUT (Fig. 3h). Unlike loss of VST-1, loss of EAT-4/VGLUT did not cause a measurable change in vesicular pH. The effect of VST-1 mutation on vesicular pH provides additional evidence that VST-1 functions in the synaptic vesicle membrane. These data are also consistent with a model in which VST-1 is an anion transporter that competes with EAT-4/VGLUT for the electrochemical gradient required for glutamate uptake into synaptic vesicles. However, some SLC17 family transporters can cotransport cations such as Na+ and H+ 33, and we cannot rule out the possibility that cation efflux (rather than anion influx) contributes to the effect of VST-1 on vesicular pH."
K	Choi *et al.* 2021	"We further tested whether the effects of vst-1 mutation on RIA activation by BAGs require GLR-1 glutamate receptors, as predicted by our model. In mutants lacking GLR-1, there was no clear effect of vst-1 mutation (Fig. 6d, e), indicating that the increased activation of RIAs observed in vst-1 mutants requires signaling through GLR-1."
L	Choi *et al.* 2021	"To determine whether increased activation of RIA contributes to the chemotaxis defects of vst-1 mutants, we tested the effect of vst-1 mutation on CO2 avoidance in animals lacking RIA interneurons. As we observed previously, vst-1 mutants displayed a robust chemotaxis defect (Fig. 6f). Animals lacking RIAs in a wild-type background did not display a clear defect in chemotaxis behavior (Fig. 6f). Notably, loss of RIAs in vst-1 mutants restored CO2 chemotaxis (Fig. 6f). These data indicate that the chemotaxis defect of vst-1 mutants is caused, at least in part, by ectopic activation of RIA interneurons."

826   **Table 5: Author Statements Collection B**

	Author	Author Statement
A	Bretscher *et al.* 2011	"When placed in a 5%-0% CO2 gradient, *C. elegans* migrate away from high CO2 (Figures 1A and 1B) (Bretscher *et al.*, 2008). We used this assay to identify potential CO2-sensing neurons..... We next attempted to rescue the tax-2(p694) defect by expressing tax-2 cDNA from neuron-specific promoters, confirming appropriate expression by polycistronic constructs that coexpress tax-2 and gfp (Coates and de Bono, 2002). Expressing tax-2 cDNA in the AFD thermosensory neurons strongly rescued CO2 avoidance, both on and off food (Figure 1D). In contrast, restoring tax-2 to the BAG O2-sensing neurons rescued CO2 avoidance on food, as shown previously (Hallem and Sternberg, 2008), but not off food. Expressing tax-2 cDNA in the ASE taste neurons or in the AQR, PQR, and URX O2-sensing neurons also partially rescued CO2 avoidance, both on food and off food (Figure 1D). These data implicate functionally diverse sensory neurons in CO2 avoidance.""
B	Kopchock *et al.* 2021	"Optogenetic stimulation of the vulval muscles triggered an immediate rise in vulval muscle cytosolic Ca2+, tonic contraction of the vulval muscles, vulval opening, and egg release (Fig. 7B,C). Although optogenetic stimulation resulted in sustained vulval muscle Ca2+ activity and contraction, vulval opening and egg release remained rhythmic and phased with locomotion, as previously observed in WT animals (Collins and Koelle, 2013; Collins *et al.*, 2016). Simultaneous brightfield recordings showed the vulva only opened for egg release when the adjacent ventral body wall muscles were in a relaxed phase (Movie 5). We have previously shown that eggs are preferentially released when the vulva is at a particular phase of the body bend, typically as the ventral body wall muscles anterior to the vulva go into a more relaxed state (Collins and Koelle, 2013; Collins *et al.*, 2016). We now interpret this phasing of egg release with locomotion as evidence that vulval muscle Ca2+ activity drives contraction, but the vulva only opens for successful egg release when contraction is initiated during relaxation of the adjacent body wall muscles. Together, these results show that optogenetic stimulation of the vulval muscles is sufficient to induce vulval muscle Ca2+ activity for egg release in a locomotion phase-dependent manner."
C	Branicky *et al.* 2016	"Because clh-3 encodes chloride channels, we reasoned that it might affect HSN activity by affecting HSN excitability. To test this, we crossed the clh-3 mutants with an integrated transgenic line that expresses Channelrhodopsin-2 (ChR2), the blue-light- activated cation channel (Nagel *et al.*, 2005), in the HSNs (wzIs6 [pegl-6::ChR2]; Leifer *et al.*, 2011; Emtage *et al.*, 2012). In wild- type worms, egg laying is robustly stimulated by ChR2 activation (Leifer *et al.*, 2011; Fig. 7). The magnitude of the response, as indicated by both the percentage of stimulations resulting in egg- laying events and the number of eggs laid per stimulation, is dependent on both the strength and duration of the light stimu- lus (Fig. 7A). The response is also completely dependent on the addition of all-trans retinal, the cofactor for ChR2, to the plates (Fig. 7B), as well as the presence of the HSNs (Fig. 7D). We observed that the clh-3(n995gf) mutant laid significantly fewer eggs per stimulation than the wild-type and blue light stimulation elicited an egg-laying event significantly less frequently in mutant animals than in wild-type. Conversely, the clh-3(ok768 and ok763) mutants laid significantly more eggs than the wild-type and blue light stimulation elicited egg-laying events, including the laying of multiple eggs, more frequently than for the wild- type (Fig. 7C,D). Together, these data support a role for the clh- 3-encoded channels in inhibiting HSN excitability: increased channel activity inhibits HSN excitability whereas loss of the channel promotes HSN excitability."
D	Waggoner *et al.* 2000a	"To assess the possible role of tpa-1 in nicotine adaptation, we first assayed the effect of tpa-1 mutations on the induction of egg-laying by levamisole and on the inactivation of this response by long-term nicotine treatment. We found that naive tpa-1 mutant animals laid eggs in response to levamisole over the same range of concentrations as did wild-type, although the magnitude of this stimulation was somewhat lower in the tpa- 1 mutant (Fig. 6A). Strikingly, however, we observed that tpa-1 mutants showed little or no reduction in levamisole-induced egg-laying after overnight exposure to nicotine (Fig. 6B), indicating that tpa-1 mutant animals were completely defective in nicotine adaptation. "
E	Waggoner *et al.* 2000a	""We also tested the effect of tpa-1 on the ability of nicotine to decrease UNC-29 receptor abundance in the vulval muscles. We observed that tpa-1 mutant animals expressing the UNC-29::GFP protein chimera in the vulval muscles retained high levels of fluorescence even after overnight nicotine treatment (Fig. 6C,D). Thus, the tpa-1-encoded isoform of PKC appeared to be necessary for the nicotine-induced downregulation of UNC-29 receptor abundance and biological activity in the vulval muscles."

F	Emtage *et al.* 2012	"Having established a method for exciting the HSN neurons in freely behaving animals, we next tested whether Go signaling controls the sensitivity of the HSNs to ChR2-mediated stimulation. egl-10 encodes an RGS family GTPase-activating protein (GAP) that accelerates hydrolysis of GTP by Goα and thereby antagonizes Go signaling (Koelle and Horvitz, 1996). egl-10 mutants carrying a Promegl-6::ChR2 transgene did not lay eggs in response to a photostimulus that reliably evoked egg-laying behavior when applied to wild-type transgenic animals (Fig. 6E), indicating that globally increasing Go signaling reduced the excitability of the HSN neurons. We next measured the effect of activating Go signaling downstream of the EGL-6 GPCR by testing the behavioral re- sponses of transgenic egl-6(gf) mutants to photostimulation. Like egl-10 mutants, transgenic egl-6(gf) mutants had reduced behavioral responses to photostimulation of HSN neurons (Fig. 6F). Deletion of irk-1 significantly restored the response of egl-6(gf) mutants to excitatory input (Fig. 6F). "
G	Brundate *et al.* 1996	"Egg laying is mediated in part by serotonin released from the HSN neurons, which signals the vulva and uterine muscles to contract. Egg-laying deficient mutants have been categorized as defective in generating the serotonergic signal (i.e., defective in the HSN), or defective in responding to the signal (defective in vulva or uterine muscles) based on their ability to lay eggs when bathed in serotonin (Trent *et al.* 1983). All of the egl-30 mutants (except n715n1190) showed a reduced egg-laying response to exogenous serotonin (Figure 5A; Trent *et al.* 1983); the responses of the mutants were inversely proportional to the severity of their egg-laying defects under growth conditions (Figure 3C). The severe egl-30 mutant ad805 laid fewer than 5% of the wild-type number of eggs in serotonin, even though animals missing HSNs lay a similar number of eggs to wild type in this assay (Trent *et al.* 1983). "
H	Collins *et al.* 2016	"We have previously shown that two Cl- extruding transporters, KCC-2 and ABTS-1, are expressed in the HSNs where they promote the development of inhibitory ligand-gated Cl- channel signaling (Tanis *et al.*, 2009; Bellemer *et al.*, 2011). These data suggest that tyramine signaling through LGC-55 would hyperpolarize the HSN and inhibit activity. To test this directly, we compared HSN activity in wild-type and lgc-55 mutant animals. We observed a significant increase in the frequency of Ca2+ transients in HSNs of lgc-55 mutant animals (Figure 6E and F) in both the inactive and active states of egg-laying behavior. Mean HSN inter-transient intervals in wild-type animals were 41 ± 5 s in the inactive state and 17 ± 2 s during the active state, while intervals in lgc-55 mutants were reduced to 22 ± 2 s in the inactive state and 13 ± 1 s during the active state. Thus, the absence of inhibitory feedback by tyramine signaling onto the HSNs leads to increased activity in both the active and inactive egg-laying behavior states."
I	Waggoner *et al.* 1998	"The roles of individual neurons in controlling the timing of egg-laying events can be determined with high precision by eliminating specific neurons by laser ablation and assaying the effect of the ablation on behavior. We therefore eliminated the neurons with prominent synaptic input to the egg-laying muscles to determine how their absence affected the timing of egg-laying events. We first investigated the involvement of the HSNs, a pair of serotonergic motor neurons that are required for efficient egg laying. By tracking the behavior of animals lacking both HSNs, we found that elimination of the HSNs did not qualitatively alter the pattern of egg laying: eggs were still laid in clusters, and the intervals between clusters and between egg-laying events within a cluster were still exponentially distributed. However, HSN ablation did cause a substantial lengthening of the inactive phase, which led to a slower overall rate of egg laying (Figure 2A). Since loss of the HSNs decreased the frequency of egg-laying clusters (i.e., λ2 was decreased; Table 1) but did not slow the egg-laying rate within these clusters (λ1 was actually increased), these results suggest that the HSNs stimulate egg laying by inducing the active state."
K	Bretscher *et al.* 2011	"The timing of CO2-evoked Ca2+ responses in both AFD and BAG correlated with peaks in locomotory activity (Figure 6A). We investigated these correlations directly by ablating AFD and/or BAG and examining behavioral responses (Figure 6B). For statistical comparison, we chose time intervals before and after gas switches according to the occurrence of peaks in wild-type behavioral rates. In the absence of food, neither AFD nor BAG ablation abolished modulation of speed across shifts in CO2 (Figures 6B and S4). Stronger phenotypes were observed for reversal and omega rates (Figure 6B). Unexpectedly, ablation of AFD increased reversal and omega rates following a sharp CO2 rise (ttx-1, Figures Figures6B,6B, B,7B,7B, 7C, 7H, and 7I) and reduced suppression of omega turns following a CO2 fall (ttx-1, Figures Figures6B,6B, B,7K,7K, and 7L), suggesting that AFD acts to suppress reversals and omega turns at these two time points. Ablation of BAG abolished reversal and omega responses to a rise in CO2 (pBAG::egl-1, Figures Figures6B,6B, B,7B,7B, 7C, 7H, and 7I) and reduced the suppression of omega turns following a CO2 fall (pBAG::egl-1, Figures Figures6B,6B, B,7K,7K, and 7L), consistent with BAG excitation promoting reversals and omega turns. Coablation of AFD and BAG abolished the suppression of reversals and omega turns following a fall in CO2 (ttx-1; pBAG::egl-1, Figures 7F and 7L). This effect was due to reduced reversal and omega rates under prolonged high CO2 (ttx-1; pBAG::egl-1, red bars, Figures 7E and 7K). These data suggest that together BAG and AFD act to suppress reversals and omega turns when CO2 decreases."

| J | Shyn *et al.* 2003 | "Behavioral data implicated serotonin, a neuromodulator released from the HSN egg-laying motorneurons, in the control of egg-laying behavior 2, 3, 4. When we treated animals with exogenous 5HT, we observed a significant increase in the frequency of Ca2+ events from a baseline of 5.63 min–1 to a rate of 35.01 min–1 (p < 0.001, Kolmogorov-Smirnov test). ...Thus, exogenous serotonin appeared to modulate the functional state of the vulval muscles, switching them from a pattern of sporadic Ca2+ activity to a pattern of continual Ca2+ activity. In principle, serotonin could exert its effects directly on the vulval muscles, or it could act indirectly by altering the activity of the egg-laying motorneurons. To resolve this issue, we ablated the egg-laying motorneurons and assayed the effect of serotonin on vulval muscle Ca2+ transients. We found that ablated animals exhibited a continuous train of Ca2+ transients on serotonin essentially identical to that exhibited by unablated wild-type animals (Figure 2, Table 1). Thus, the ability of serotonin to increase the frequency of Ca2+ events was not markedly affected by the absence of the egg-laying motorneurons, indicating that serotonin directly stimulates the activity of the vulval muscles." |

Figure Titles and Captions

Figure 1 **Standard GO annotations and GO-CAMs A)** Standard GO annotations link genes (pink text) to GO Molecular Functions
(white text) GO Biological Processes (dark red text) or GO Anatomy terms (blue box). **B)** A partial GO-CAM of canonical glycolysis.
Gene Ontology-Causal Activity Models (GO-CAMs) arrange GO annotations into structured models of biological processes by
causally linking GO molecular functions that make up a process. Edges represent relations which may connect nodes according to
the GO-CAM data model. **C)** A key describing the correspondence of box color to the parent classes in the ontologies used in *Ce*N-
CAM.

*Figure 2 Ce*N-CAM **annotations link cells to behaviors**. *Ce*N-CAM annotations link cells to behaviors. Cells such as HSN are
represented with unique identifiers in the *C. elegans* Gross Anatomy Ontology. **A)** Cell ablation phenotypes can be modelled using
the generic GO *cellular process* class to reflect the non-molecular nature of the experiment, and *causally upstream of or within*
relations, allowing for the most inclusive description of the relationship between *cellular process* and *nervous system process*
terms. Example drawn from Waggoner et al. (1998) (Table 4A, this manuscript).   **B)** A *Ce*N-CAM model describing the role of the
acetylcholine biosynthetic process in the VC neuron in the regulation of egg-laying (Bany *et al.* 2003) (Table 4B, this manuscript).
Because the *acetylcholine biosynthetic process* can proceed independently of electrical activity, it is modelled as *causally upstream*
of, positive effect; acetylcholine secretion, neurotransmission. **C)** Optogenetic activation of uv1 leads to a decrease in egg-laying
(Banerjee *et al.* 2017) (Table 4C, this manuscript). This *membrane depolarization* process is modelled as part of the *negative*
regulation of egg deposition, a *nervous system process*.

*Figure 3 Ce*N-CAM **models of inputs to neurons and behavior.** Cell-specific genetic rescue of a behavioral response to
pharmacological treatment or environmental stimuli produces models linking genes, GO molecular functions, GO biological
processeses, and cells in the *C. elegans* Gross Anatomy Ontology. **A)** Carnell *et al.* (2005) (Table 4D, this manuscript) found that VM-
specific expression of *ser-1* could rescue serotonin-dependent egg-laying behavior, suggesting that *ser-1* is required in VM neurons
to induce egg-laying in response to serotonin. The *G-protein coupled serotonin receptor activity* is *part of* the *positive regulation of*
egg deposition, because the *part of* relation is transitive (i.e. there is no need for an additional *part of* relation connecting these
nodes). **B)** Carillo *et al.* (2013) (Table 4E, this manuscript) found that *npr-1* expression in neuronal subsets that include URX is
sufficient to rescue the *behavioral response to carbon dioxide*. The activity of some CO_2 receptor is implied, leading to the addition
of a placeholder term without an enabling gene   , indicating an important piece of missing information. This activity can be
included in the CO_2 sensing circuit by asserting that it is *part of* the *positive regulation of chemotaxis*, along with the *npr-1*-
dependent signaling pathway. **C)** AFD neuron responds to CO_2 removal (Bretscher *et al.* 2011) (Table 4F, this manuscript). Currently,
there are no terms within appropriate ontologies to describe temporal features of chemical or physical inputs (e.g. 'decreasing').
The required definitions are suggested in this paper.

*Figure 4 Ce*N-CAM **models of neuron-to-neuron functional connectivity** **A)** Optogenetic activation of HSN neuron causes
membrane depolarization in HSN (Collins *et al.* 2016) (Table 4G, this manuscript). **B)** Inhibiting neuron-neuron synaptic
transmission in VC causes increased activity (membrane depolarization) in HSN, suggesting inhibition of HSN dependent on synaptic
transmission from VC (Kopchock *et al.* 2021) (Table 4H, this manuscript). **C)** Model of BAG-RIA activating connection based on data
from Choi *et al.* (2021). Blue boxes represent a GO cellular component term.

*Figure 5 Ce*N-CAM **representation of several cells in the egg-laying circuit and their interactions.** Drawn from several statements
in Supplementary Table 1 (Carnell *et al.* 2005; Collins *et al.* 2016; Fenk & de Bono 2015; Kim *et al.* 2001; Kopchock *et al.* 2021).
Compared with models of individual author statements, more specific relations can be used here, given the biological context (for
instance *positively regulates*, rather than *causally upstream of, positive effect*).

Figure 6 **Modelling signal integration with the *part of* relation.** Bretscher *et al.* (2011) (Table 5A, this manuscript) found that
restoring *tax-2* expression to BAG neurons rescued CO_2 avoidance on food, but not off food, suggesting that *tax-2*-dependent
avoidance behavior requires food input. This implies an AND-gated interaction to integrate food and CO_2 signals. **(A)** One of the

relations in GO-CAM, *causally upstream of* does not capture the necessity of each input, whereas **(B)** *part of* does imply necessity,
as required to capture the AND logic involved in sensory integration. **(C)** GO-CAM representation of the author statement listed in
Table 2A (this manuscript). **(D)** This relation may also be useful for modelling co-ordination of behaviors. Kopchock *et al.* (2021)
(Table 5B, this manuscript)　　　found that optogenetic activation of the vulval muscles was insufficient to induce an egg-laying
event; instead co-ordination of VM activation with a particular phase in the body bend during locomotion was required.

*Figure 7 Ce*N-CAM model of sensory adaptation to acetylcholine Chronic exposure to achetylcholine leads to *tpa-1*-dependent
downregulation of *unc-29*. This reduces acetylcholine receptor activity (Waggoner *et al.* 2000a) (Table 4D, 4E, this manuscript).
Duplicating the terms for acetylcholine receptor activity and the associated signaling pathway enables the model to distinguish
between different temporal instances of molecular functions or processes, where the input requirements are different in each
case. PATO terms specifying duration and relative amount of input are necessary to fully represent sensory adaptation.

Figure 8 Proposed classes for addition to the Environmental Conditions, Treatments & Exposures Ontology (ECTO) to represent
temporal features of environmental inputs. The new classes *increasing amount* and *decreasing amount* can be used in combination
with the relation *has quality* (RO:0000086).

Figure 9 **Functional annotation of the *C. elegans* connectome allows visualization of causal relationships within and among**
neural circuits. Black or grey arrows indicate synaptic connections from electron microscopy of serial sections. Coloured solid
arrows indicate activating (green) or inhibiting (blue) synaptic connections. Coloured dotted arrows indicate activating or inhibiting
indirect or extra-synaptic connections. Bold outlines on neurons indicate the sign of the neuron on the specified behavior. Fill
colour on neurons indicates the effect of the specified input on neural activity. **A)** Synaptic connectivity for subset of neurons
involved in both CO_2 avoidance behavior and egg-laying behavior. **B)** Neurons activated or inhibited by CO_2 are indicated by fill
color, neurons contributing to CO_2 avoidance behavior indicated by outline color. **C)** Neurons activated or inhibited by serotonin are
indicated by fill color, neurons contributing to egg-　　laying behavior indicated by outline color.

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Key

[WB Gene ID]
Gene

[GO ID]
GO Molecular Function

[GO ID]
GO Cellular Component

[GO ID]
GO Biological Process

[CHEBI ID]
chemical

[WB Anatomy ID]
Cell Type